



## RESERVATIONS AND TIMETABLING



### Service scheduling characteristics

- ❑ Operations → **activities**
  - e.g. meetings to be attended by certain people, game to be played by two teams.
- ❑ Data:
  - Processing time → duration
  - Release time → earliest possible start time
  - Due date → latest possible end time
  - Weight → priority

João Miguel da Costa Sousa

381



### Service scheduling characteristics

- ❑ **Resources:**
  - Classroom, hotel, rental car, stadium, operating room, plane, ship, airport gate, dock, railroad track, person (nurse/pilot)
- ❑ Synchronization of resources may be important
  - Need a plane and a pilot
  - Classroom, AV equipment, professor, students
- ❑ Each resource may have its own characteristics
  - **Classroom:** capacity, equipment, cost, accessibility
  - **Truck:** capacity, refrigeration, speed
  - **Person:** specialist (surgeon, nurse) with skills (languages)

João Miguel da Costa Sousa

382



### Service scheduling characteristics

- ❑ **Activities** may have
  - Time windows
  - Capacity requirements/constraints
- ❑ **Resources** may have
  - Setup/transition time – runways at airports
  - Operator/tooling requirements
  - Workforce scheduling constraints
    - Shift patterns, break requirements
    - Union and safety rules

João Miguel da Costa Sousa

383



### Differences from manufacturing

- ❑ Impossible to “store” goods
  - If a hotel room is not filled it is not possible to “get back” the lost time
- ❑ Resource availability often varies
  - May even be part of the objective function
- ❑ Saying “no” to a customer is common
  - “No available seats on that flight” (even if there are some!)
  - Try to book a restaurant for 8 PM

João Miguel da Costa Sousa

384



### Reservation systems and timetabling

- ❑ Hotel rooms, car rentals, airline tickets (and classroom scheduling)
- ❑ Utilization of a resource for a given period of time
  - With slack:  $p_j \leq d_j - r_j$
  - Without slack  $p_j = d_j - r_j$
- ❑ May not be able to schedule all requests

João Miguel da Costa Sousa

385



## Objectives

- ❑ Maximize profit
- ❑ Maximize resource usage
- ❑ Minimize number of rejected requests
- ❑ Minimize losses of rejected requests



## Reservation systems and timetabling

1. Reservation Systems without Slack (timing of job fixed)
  - $m$  parallel machines (resources),  $n$  jobs (activities) fixed in time; interval scheduling
2. Reservation Systems with Slack
  - $m$  parallel machines,  $n$  jobs (more flexible in time)
3. Timetabling with Workforce Constraints
  - $W$  identical resources in parallel,  $n$  jobs using one resource
4. Timetabling with Operator or Tooling Constraints
  - Can require more than one resource that are *not* identical.



## Reservation Systems without Slack

- ❑ Reservation system with  $n$  activities,  $m$  resources
  - ❑ Release date  $r_j$ , due date  $d_j$ ,
 

Weight is often equivalent to profit
  - ❑ Activities have weight  $w_j$  or  $w_{ij}$
  - ❑ All activities and resources are independent
  - ❑ No slack:  $p_j = d_j - r_j$ . If accepted, job starts at time  $r_j$ .
  - ❑ Can we accept all activities? Do we need all resources?
- Objective 1:** Maximize number of activities processed    **OR**
- Objective 2:** Min. number of resources needed for all activities



## Example: car rental

- ❑ Four types of cars: subcompact, midsize, full size, and sport utility
- ❑ Fixed number of each
- ❑ Resource = type of car
- ❑ Activity = customer requesting a car
- ❑ May have resource subsets
  - Rent a subcompact or midsize car
  - Some substitutability of resources



## Example: car rental

- ❑ Reservation is for  $p_j$  days.
- ❑ Profit for car type  $i$  is  $\pi_i$ . ( $\pi_{ij}$  if it depends on customer)
- ❑ Weight  $w_j$
- ❑ Time is divided into  $H$  (integer) slots.
- ❑  $x_{ij}$ : binary variable that is 1 if activity  $j$  is assigned to resource  $i$
- ❑  $J_t$ : set of activities that need a resource in slot  $t$  [ $t-1, t$ ]



## Feasibility problem

- ❑ Can we process every activity?
  - ❑ Can we assign activities to resources such that activity  $j$  is assigned to a set  $M_j$  of resources?
- Relatively easy to solve



## Problem formulation

### Integer Programming problem

$$\max \sum_{i=1}^m \sum_{j=1}^n w_{ij} x_{ij}$$

Maximize total profit

$$\sum_{i=1}^m x_{ij} \leq 1, \quad j = 1, \dots, n$$

Every activity is assigned to at most one resource

$$\sum_{j \in J_t} x_{ij} \leq 1, \quad i = 1, \dots, n, \quad t = 1, \dots, H$$

Each resource has only one activity per time slot

João Miguel da Costa Sousa

392



## IP formulation

□ In general the problem is NP-hard (when both  $p_j$  and  $w_{ij}$  are free)

➤ Two special cases with exact algorithms

- each of these algorithms sorts the activities to increasing  $r_j$

□ **Case 1: processing times  $p_j = 1$**

- decompose into separate time units
- assign at each time unit most valuable activities first
- **it is an independent problem for each time slot!**

João Miguel da Costa Sousa

393



## 2: Identical weights and machines

□ Assume  $w_j = 1$  and  $M_j = \{1, \dots, m\}$  (all resources identical)

□ **Objective:** maximize number of activities assigned

➤ No time decomposition, but still an exact, simple algorithm can be applied:

- Order activities in increasing order of release dates:

$$r_1 \leq r_2 \leq \dots \leq r_n$$

- Let  $J$  be the set of already scheduled activities

João Miguel da Costa Sousa

394



## Algorithm: max. activities assigned

**Step 1:**

□ Set  $J = \emptyset$  and  $j = 1$

**Step 2**

□ If a resource is available at time  $r_j$ , assign it to activity  $j$ , add activity  $j$  to  $J$ , and go to Step 4.

□ Otherwise go to Step 3.

**Step 3**

□ Let  $j^*$  be such that (maximum completion time):

$$C_{j^*} = \max_{k \in J} (C_k) = \max_{k \in J} (r_k + p_k)$$

João Miguel da Costa Sousa

395



## Algorithm: max. activities assigned

□ If  $C_j = r_j + p_j > C_{j^*}$ , do not include activity  $j$  in  $J$  and go to Step 4.

□ Else, delete activity  $j^*$  from  $J$ , assign activity  $j$  to the resource freed and include activity  $j$  in  $J$

**Step 4**

□ If  $j = n$  STOP,

□ Else, set  $j = j + 1$  and return to Step 2

João Miguel da Costa Sousa

396



## Unlimited # of resources

□ No slack, arbitrary processing times, equal weights, identical resources

□ Infinitely many resources in parallel

➤ **Minimize the number of resources used**

□ Easily solved

□ Order jobs as before

João Miguel da Costa Sousa

397



## Algorithm

- ❑ Assign activity 1 to resource 1
- ❑ Suppose first  $j - 1$  activities have been processed
- ❑ Try to assign  $j^{\text{th}}$  activity to a resource in use
- ❑ If not possible assign to a new resource

➤ *Special case of the node coloring problem in graph theory.*



## Node coloring problem

- ❑ Consider a graph with  $n$  nodes
- ❑ If an arc  $(j, k)$  connects nodes  $j$  and  $k$  they cannot be colored with the same color
- ❑ How many colors do we need to color the graph?

**Number of colors needed = Number of resources needed**



## Reservation systems with slack

- ❑ Now allow slack  $p_j \leq d_j - r_j$  (time slots)
- Trivial case
  - all processing times  $p_j = 1$ , identical weights, identical machines
  - Schedule constructed progressively in time
- Non-identical processing times, weights  $w_j$ 
  - NP-hard →
  - No efficient algorithm →
  - Heuristic needed!



## Composite dispatching rule

- ❑ Preprocessing: determine flexibility of activities and resources
- ❑ Dispatch least flexible activity first on the least flexible resource, etc.
  - $v_{it}$  is the number of activities that may be assigned to resource  $i$  during interval  $[t-1, t]$ .
  - $|M_j|$  is the number of resources in set  $M_j$  (suitable for activity  $j$ ).



## Priority indices

- ❑ Priority index for **activities**

$$I_j = f(w_j / p_j, |M_j|)$$

- The higher  $w_j/p_j$  and the smaller  $|M_j|$ , the lower the index.
- Activities can be ordered in decreasing order:  $I_1 \leq I_2 \leq \dots \leq I_n$

- ❑ Priority indexes for **resources** (examples for  $[t, t+p_j]$ )

$$g(v_{i,t+1}, v_{i,t+2}, \dots, v_{i,t+p_j}) = \sum_{l=1}^{p_j} v_{i,t+l} / p_j \quad \text{or}$$

$$g(v_{i,t+1}, v_{i,t+2}, \dots, v_{i,t+p_j}) = \max(v_{i,t+1}, v_{i,t+2}, \dots, v_{i,t+p_j})$$



## Algorithm

### Step 0

- ❑ Calculate both priority indices
- ❑ Order activities according to activity priority index  $I_j$

### Step 1

- ❑ Set  $j = 1$ .

### Step 2

- ❑ For activity  $j$  select the resource and time slot with lowest resource index  $g(v_{i,t+1}, v_{i,t+2}, \dots, v_{i,t+p_j})$
- ❑ Discard activity  $j$  if it cannot be assigned to any resource

### Step 3

If  $j = n$  STOP; otherwise set  $j = j+1$  and return to Step 2



### Example 9.3.2

Consider seven activities and three resources:

Activities	1	2	3	4	5	6	7
$p_j$	3	10	9	4	6	5	3
$w_j$	2	3	3	2	1	2	3
$r_j$	5	0	2	3	2	4	5
$d_j$	12	10	20	15	18	19	14
$M_j$	{1,3}	{1,2}	{1,2,3}	{2,3}	{1}	{1}	{1,2}

João Miguel da Costa Sousa

404



### Example 9.3.2 (cont.)

Consider:

$$I_j = f(w_j / p_j, |M_j|) = \frac{|M_j|}{w_j / p_j}$$

The indices for the activities are the following:

Activities	1	2	3	4	5	6	7
$I_j$	3	6.67	9	4	6	2.5	2

João Miguel da Costa Sousa

405



### Example 9.3.2 (cont.)

Factors  $v_{it}$  are:

slot t	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19
$v_{1t}$	1	1	3	3	4	6	6	6	6	6	5	5	4	4	3	3	3	3	2	1
$v_{2t}$	1	1	2	3	3	4	4	4	4	4	3	3	3	3	2	1	1	1	1	1
$v_{3t}$	1	0	1	2	2	3	3	3	3	3	3	3	2	2	2	1	1	1	1	1

Applying

$$g(v_{i,t+1}, v_{i,t+2}, \dots, v_{i,t+p_j}) = \sum_{l=1}^{p_j} v_{i,t+l} / p_j$$

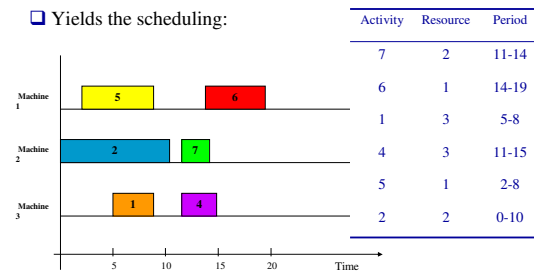
João Miguel da Costa Sousa

406



### Example 9.3.2 (cont.)

Yields the scheduling:



João Miguel da Costa Sousa

407



### Example 9.3.2: discussion

- Activity 3 does not fit into the schedule.
- Optimal schedule:
  - activity 7 starts with resource 1 at time 10
  - activity 3 starts with resource 2 at time 11
- Heuristic yields a suboptimal solution.
- Other functions for the priority indices for activities yield different schedules (see Pinedo's book).

João Miguel da Costa Sousa

408



### Timetabling with Workforce Constraints

- Infinite identical resources in parallel
- $n$  activities and all have to be done
  - Processing time of activity  $j$  is  $p_j$
  - No preemption
- Total number of identical operators is  $W$
- Each activity requires  $W_j$  operators
- Clearly, if  $W_j + W_k > W$  then activity  $j$  and activity  $k$  cannot be processed at the same time.

João Miguel da Costa Sousa

409



## Timetabling with Workforce Constraints

- Find a schedule that minimizes makespan
- ❑ Special case of **Project Scheduling with Workforce Constraints** with one resource,  $p_j \geq 1$ , no precedence

### ❖ Applications

- scheduling a construction project ( $W$  is the crew size)
- exam scheduling ( $W$  is the number of seats)

João Miguel da Costa Sousa

410



## Example 9.4.2: Exam scheduling

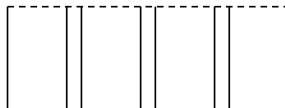
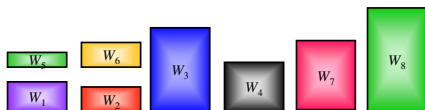
- ❑ All exams have the same duration
- ❑ One exam room with capacity  $W$
- ❑ Course  $j$  has  $W_j$  students
- ❑ All students in course  $j$  must take the exam at the same time
- ❑ Find a timetable for all  $n$  exams in the minimum amount of time

João Miguel da Costa Sousa

411



## Special case: bin packing



Pack the objects into the bins to minimize the number of bins used

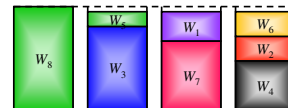
Each bin has capacity  $W$

João Miguel da Costa Sousa

412



## Special case: bin packing



Pack the objects into the bins to minimize the number of bins used

Each bin has capacity  $W$

João Miguel da Costa Sousa

413



## Bin packing problem

- ❑ Assume equal processing times  $p_j = 1$
- ❑ Unlimited number of machines
- ❑ Minimize makespan
- ❑ **Equivalent to bin packing problem**
  - each bin has capacity  $W$  and is one time slot
  - activity is an item of size  $W_j$
  - items packed in one bin are activities done in time slot
  - pack into the minimum number of bins

João Miguel da Costa Sousa

414



## Solving the bin packing problem

- ❑ Known to be NP-hard
- ❑ Many heuristics developed
- ❑ **First Fit (FF) heuristic**
  - Order activities (items) in an arbitrary order
  - Always put an item in the first bin it fits into
  - Known that

$$C_{\max}(\text{FF}) \leq \frac{17}{10} C_{\max}(\text{OPT}) + 2$$

João Miguel da Costa Sousa

415



### Example 9.4.3

- Assume 18 activities and  $W = 2100$ .

activities	1,...,6	7,...,12	13,...,18
$W_j$	301	701	1051

**FF heuristic:**

- Assign the first 6 activities to one interval ( $301 \times 6 = 1806$ )
- Then assign two activities of 7,...,12 at a time to the next 3 intervals ( $701 \times 2 = 1402$ )
- Finally, assign one activity of 13,...,18 to each interval.

João Miguel da Costa Sousa

416



### Example 9.4.3

- Makespan of FF:  $C_{\max} = 10$ .

➤ *Poor performance when jobs are assigned in arbitrary order!*

- Optimal solution:** assigns to each slot of time three activities: one of 301, one of 701 and one of 1051 ( $C_{\max} = 6$ ).

João Miguel da Costa Sousa

417



### First Fit Decreasing (FFD)

- An improvement of FF.
- Order activities in decreasing order of  $W_j$ .

- Known that  $C_{\max}(\text{FFD}) \leq \frac{11}{9} C_{\max}(\text{OPT}) + 4$

❖ Find solution of Example 9.4.3 using FFD.

- FF and FFD can be extended to activities with different release dates.**

João Miguel da Costa Sousa

418



### Example 9.4.4

- 30 activities and  $W = 1000$ .

activities	1,...,6	7,...,12	13,...,18	19,...,30
$W_j$	501	252	251	248

**Optimal schedule:**

- assign to each of the first 6 slots 3 activities: one of 501, one of 251 and one of 248.
- To the remaining 3 slots it assigns 4 activities: two of 252 and two of 248  $\rightarrow C_{\max} = 9$ .

João Miguel da Costa Sousa

419



### Example 9.4.4

**FFD rule:**

- assign to each of the first 6 slots 2 activities: one of 501 and one of 252.
- To the next 2 slots it assigns three activities of 252.
- To the remaining 3 slots it assigns four activities of 248  $\rightarrow C_{\max} = 11$ .

João Miguel da Costa Sousa

420



### Timetabling with Operator Constraints

- Up to now, the  $W$  operators are equivalent

- Any set of  $W_j$  operators could do activity  $j$

- This is not true in many applications

- e.g., medical or language specialties

➤ **Operator or Tool Constraints**

- an operator need specific skills to do an activity
- a specific tool is needed

- If two activities require the same operator or tool, they cannot be done at the same time.

João Miguel da Costa Sousa

421



## Timetabling with Operator Constraints

- ❑ All  $n$  activities must be processed
- ❑ Minimizing makespan
- ❑ Timetabling is a special case of the *project scheduling with workforce constraints* problem:
  - no precedence constraints
  - One operator per skill (or one tool)  $w_i = 1$ .
- ❑ NP-hard
- ❑ Assume all activities durations = 1
- ❑ Equivalent to the **node (graph) coloring problem**

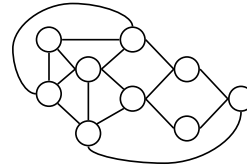
João Miguel da Costa Sousa

422



## Node Coloring Problem

- ❑ Consider a graph with  $n$  nodes that are to be colored
- ❑ An arc  $(j,k)$  **cannot** receive same color.
- ❑ Color each node such that
  - No two connected nodes have the same color
  - Use the minimum number of colors



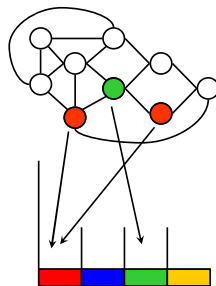
João Miguel da Costa Sousa

423



## Node Coloring Problem

- ❑ Nodes are activities
- ❑ Arcs mean the activities require the same operator
- ❑ Colors are time slots
  - Minimizing the number of colors is minimizing makespan



João Miguel da Costa Sousa

424



## Node Coloring Problem

- ❑ Recall
  - Activity = node
  - Activities need same tool = arc between nodes
- ❑ **Feasibility problem:**
  - Can the graph be colored with  $m$  (or less) colors?
- ❑ **Optimality problem:**
  - What is the lowest number of colors needed?

João Miguel da Costa Sousa

425



## Heuristics for Node Coloring

- ❑ Terminology
  - **degree of node** = number of arcs connected to node
  - **saturation level** = number of colored nodes connected to node (in a partially colored graph)
- Many heuristics exist for graph coloring
- ❑ Intuition
  - Color high degree nodes first
  - Color high saturation level nodes first

João Miguel da Costa Sousa

426



## Algorithm for Node Coloring

- ❑ **Step 1:** Order nodes in **decreasing order of their degree**.
- ❑ **Step 2:** Color node of maximal degree with Color 1.
- ❑ **Step 3:** Choose an uncolored node with **maximum saturation level**, breaking ties according to **degree**.
- ❑ **Step 4:** Color selected node using the color with the lowest possible number.
- ❑ **Step 5:** If all nodes are colored, STOP. Otherwise go to Step 3.

João Miguel da Costa Sousa

427





### Example 9.5.2

- Four university professors have to attend several meetings.
- Goal: schedule all meetings from 2 to 6 p.m.

meetings	1	2	3	4	5	6	7
Gary	1	0	0	1	1	0	1
Hamilton	1	1	1	0	0	0	0
Izak	0	0	1	0	1	1	0
Reha	1	0	1	1	1	0	0

João Miguel da Costa Sousa

428



### Example 9.5.2

- Problem can be transformed into a timetabling problem with operator constraints.
- Activities are nodes of the graph

Activities	1	2	3	4	5	6	7
operator 1	1	0	0	1	1	0	1
operator 2	1	1	1	0	0	0	0
operator 3	0	0	1	0	1	1	0
operator 4	1	0	1	1	1	0	0

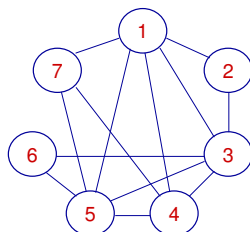
João Miguel da Costa Sousa

429



### Corresponding graph

nodes	1	2	3	4	5	6	7
degree	5	2	5	4	5	2	3



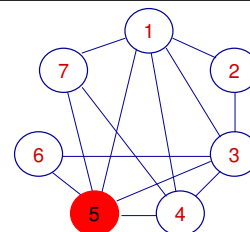
João Miguel da Costa Sousa

430



### Color first node

nodes	1	2	3	4	5	6	7
degree sub	4	2	4	3	-	1	2
saturation	1	0	1	1	-	1	1



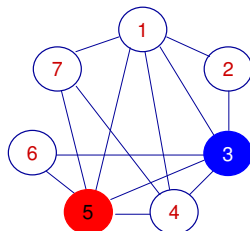
João Miguel da Costa Sousa

431



### Color second node

nodes	1	2	3	4	5	6	7
degree	3	1	-	2	-	0	2
saturation	2	1	-	2	-	2	1



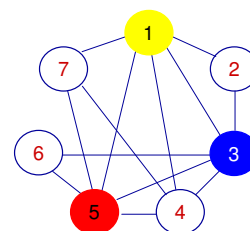
João Miguel da Costa Sousa

432



### Color third node

nodes	1	2	3	4	5	6	7
degree	-	0	-	1	-	0	1
saturation	-	2	-	3	-	2	2



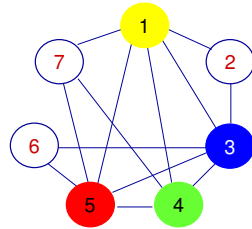
João Miguel da Costa Sousa

433



## Color rest of the nodes

nodes	1	2	3	4	5	6	7
degree	-	0	-	-	-	0	0
saturation	-	2	-	-	-	2	3



**Result 4 colors**  
7: blue  
2: red  
6: yellow

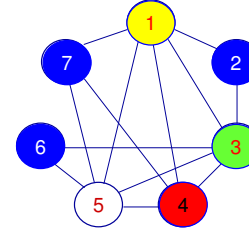
João Miguel da Costa Sousa

434



## Reserse method is worse

nodes	1	2	3	4	5	6	7
degree	5	2	5	4	5	2	3



Another color  
needed for node 5  
**Result: 5 colors**

João Miguel da Costa Sousa

435



## Relation to reservation models

- Closely related to reservation problem with zero slack and arbitrary processing times
- Special case of timetabling problem
  - tools in common = overlapping time slots
  - tools in common  $\Rightarrow$  nodes connected
  - colors = resources
  - minimizing colors = minimizing resources
  - adjacent time slots vs. tools **need not** be adjacent

João Miguel da Costa Sousa

436



## Timetabling and reservations

- Example of reservation system:

activities	1	2	3
$p_j$	2	3	1
$r_j$	0	2	3
$d_j$	2	5	4

João Miguel da Costa Sousa

437



## Equivalent timetabling problem

	activities	1	2	3
Job 1 must be processed in time [0,2]	<i>Tool 1</i>	1	0	0
	<i>Tool 2</i>	1	0	0
Job 2 must be processed in time [2,5]	<i>Tool 3</i>	0	1	0
	<i>Tool 4</i>	0	1	1
	<i>Tool 5</i>	0	1	0

João Miguel da Costa Sousa

438



## Example: classroom timetabling

- UC Berkeley
  - 30,000 students, 80 depts
  - 4000 classes, 250 rooms
  - 3 schedulers and 1 supervisor



- For each section of each course, the departments supply
  - Estimated enrollment
  - Requested meeting time
  - Special requirements (e.g., A/V equipment)

João Miguel da Costa Sousa

439



## Development of an objective

### Some obvious components:

- ☐ One class at a time in a given room, for a given prof, for a given number of students
- ☐ Usually minimize the number of students who cannot take the courses they want
- ☐ Room should be big enough
- ☐ Special equipment should be present

João Miguel da Costa Sousa

440



## Development of an objective

### Some **non-obvious** components:

- ☐ Profs want rooms close to their offices
- ☐ Students want consecutive classes to be close together
- ☐ Profs get one day with no classes
- ☐ (Departments want classes in rooms they “own”)
- ☐ (Everyone wants no classes on Friday)

João Miguel da Costa Sousa

441



## Development of an objective

- ☐ How do you balance the components?
  - Is a room within 100 m of a prof’s office worth not being able to accommodate all students?
  - “You can have a Friday afternoon class with A/V equipment or a Friday morning class without.”

João Miguel da Costa Sousa

442



## Berkeley guidelines

- ☐ Standard “calendar”
  - 9-hour day, starting at 8 AM
  - 9 1-hour blocks overlap with 6 1.5-hour blocks
- ☐ “Prime time” blocks
  - One department can only request 60% of its classes during prime time.

João Miguel da Costa Sousa

443



## Berkeley solution

- ☐ Large Integer Programming problem
  - 0.5M variables, 30K constraints
  - Very high penalty for not scheduling a class at all
  - Other objective components: distance, over-utilized facilities, empty seats
- ☐ **Solved heuristically!**
  - See Alg. 9.6.1, p 222

João Miguel da Costa Sousa

444



## Room assignment heuristic

- ☐  $J$  is the set of all classes
- ☐  $t$  is a timeslot
- ☐  $J_t$  is the set of all classes assigned to  $t$
- ☐  $M$  is the set of all classrooms
- ☐  $M_j$  is the set of classrooms that can contain class  $j$

João Miguel da Costa Sousa

445



## Room assignment heuristic

### Step 1 (*Select Time Slot*)

- ❑ Select, among time slots not yet considered, slot  $t$  with the smallest supply/demand ratio.

### Step 2 (*Greedy Algorithm*)

- ❑ Rank all classes  $j$  in  $J_t$  in decreasing order of class size.
- ❑ Assign class  $j$  to the (vacant) room  $M_j$  with lowest cost

### Step 3 (*Improvement Phase*)

- ❑ Rank all classes  $j$  in  $J_t$  in decreasing order of current cost.

João Miguel da Costa Sousa

446



## Room assignment heuristic

- ❑ If class  $j$  is not assigned, find all interchanges with a class  $k$  in an occupied room, moving  $k$  to a vacant room. Make interchange with maximum cost reduction.
- ❑ If class  $j$  is assigned, find interchanges that reduce total cost. Apply interchange with maximum cost reduction.

### Step 4 (*Stopping Criterion*)

- ❑ If Step 3 reduced the total cost, return to Step 3; otherwise update unscheduled list.
- ❑ If all time slots were selected, STOP; else go to Step 1.

João Miguel da Costa Sousa

447



## Discussion

- ❑ This chapter considered four types of scheduling and timetabling problems.
- ❑ Real world problems have all the features discussed: release dates and due dates (with or without slack), workforce and tool constraints.
- ❑ In practice:
  - Dynamic rather than static reservation systems
  - Price considerations, which depend on current occupancy and forecast demand.

João Miguel da Costa Sousa

448